

## SOJOURN IN AN ELLIPTICAL DOMAIN

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Received 6 May 1985

Revised 9 September 1985

We extend results of Maejima (1984) concerning the time that a two-dimensional stationary Gaussian process spends in an elliptical domain. Here: (a) the process may be cross-correlated while the domain is elliptical; (b) the cross-correlations do not vanish asymptotically; (c) a functional limit theorem is obtained.

self-similar process \* long-range dependence \* multiple Wiener integrals \* two-dimensional Gaussian processes \* weak convergence

### 1. Introduction

Maejima (1984) considered a two-dimensional stationary Gaussian process  $X(t) = (X_1(t), X_2(t))$  where  $X_1(t)$  and  $X_2(t)$  are correlated with long-range dependence. He obtained a non-central limit theorem for the time that the process  $X(t)$  spends in an elliptical domain  $D$  centered at the origin. We extend this result in three ways:

(1) We do not suppose that the ellipse is a circle when  $X_1(t)$  and  $X_2(t)$  are dependent.

(2) The cross-correlations between  $X_1(t)$  and  $X_2(t)$  are allowed to have the same order of magnitude as the auto-correlations.

(3) We establish convergence in  $C[0, \infty)$ , the space of continuous functions on  $[0, \infty)$  endowed with the sup-norm topology.

For a recent survey on sojourns, see Maejima (1985).

### 2. The main result

Let  $X(t) = (X_1(t), X_2(t))$ ,  $t \geq 0$ , be a two-dimensional stationary Gaussian process with

$$EX(t) = 0, \quad R(t) \equiv EX(0)^* X(t) = \begin{pmatrix} r(t) & \rho(t) \\ \rho(t) & r(t) \end{pmatrix}$$

where  $x^*$  denotes the transpose of the vector  $x$ . Suppose in addition that

Research supported by the National Science Foundation grant ECS-80-15585.

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$$\begin{aligned}
&r(t) \text{ and } \rho(t) \text{ are continuous,} \\
&r(0) = 1, \quad \rho(0) = \rho \quad (0 \leq \rho < 1), \\
&r(t) \sim t^{-\alpha} L(t) \quad \text{as } t \rightarrow \infty, \\
&\rho(t) \sim \rho_\infty t^{-\alpha} L(t) \quad \text{as } t \rightarrow \infty,
\end{aligned}$$

where  $0 < \alpha < \frac{1}{2}$ ,  $0 \leq \rho_\infty \leq 1$  and  $L$  is a slowly varying function at infinity.

The components  $X_1(t)$  and  $X_2(t)$  exhibit a long-range dependence because of the asymptotic form of  $r(t)$ . These components are said to be asymptotically independent if  $\rho_\infty = 0$ . (Maejima (1984) supposes  $\rho_\infty = 0$ .)

Let

$$D = \{(x_1, x_2): a^2 x_1^2 + b^2 x_2^2 \leq 1\}, \quad 0 < a \leq b,$$

be an elliptical domain centered at the origin with principal axes that are parallel to the coordinate axes. Let  $I[\cdot]$  be the indicator function and let

$$M(t) = \int_0^t I[X(s) \in D] ds, \quad t > 0,$$

denote the amount of time that the process  $X(\cdot)$  has spent in the domain  $D$  by time  $t$ . Our main theorem concerns the asymptotic behavior as  $t \rightarrow \infty$  of the standardized process

$$Z_t(\tau) = \frac{M(t\tau) - EM(t\tau)}{[\text{Var } M(t)]^{1/2}}, \quad \tau \geq 0.$$

$Z_t(\tau)$  depicts the deviations, adequately rescaled, of the sojourn functional  $M(\cdot)$  from its mean  $EM(\cdot)$ .

Let

$$A = \frac{1}{2}(1 + \rho)(a^2 + b^2) > 0, \quad B = (1 - \rho^2)^{1/2}(a^2 - b^2) \leq 0,$$

$$C = \frac{1}{2}(1 - \rho)(a^2 + b^2) > 0,$$

$$\alpha_1 = \left( \frac{4A}{4AC - B^2} \right)^{1/2}, \quad \alpha_2 = \left( \frac{4C}{4AC - B^2} \right)^{1/2},$$

$$\beta_1(y) = \frac{B}{2A}y, \quad \beta_2(y) = \frac{B}{2C}y,$$

$$\gamma_1(y) = \frac{1}{A^{1/2}} \left\{ \left( \frac{B^2}{4A} - C \right) y^2 + 1 \right\}^{1/2}, \quad \gamma_2(y) = \frac{1}{C^{1/2}} \left\{ \left( \frac{B^2}{4C} - A \right) y^2 + 1 \right\}^{1/2},$$

$$\phi(y) = \frac{1}{\sqrt{2\pi}} e^{-y^2/2},$$

$$\begin{aligned}
h_i(y) &= (\beta_i(y) - \gamma_i(y))\phi(\beta_i(y) - \gamma_i(y)) \\
&\quad - (\beta_i(y) + \gamma_i(y))\phi(\beta_i(y) + \gamma_i(y)) \quad \text{for } i = 1, 2,
\end{aligned}$$

$$\begin{aligned}
h_3(y) &= \exp\{-\frac{1}{2}(\beta_1(y)^2 + \gamma_1(y)^2)\} \\
&\quad \times [\exp\{-\beta_1(y)\gamma_1(y)\} - \exp\{\beta_1(y)\gamma_1(y)\}],
\end{aligned}$$

and finally let

$$c(2, 0) = \int_0^{\alpha_1} \phi(y) h_1(y) dy, \quad (2.1)$$

$$c(0, 2) = \int_0^{\alpha_2} \phi(y) h_2(y) dy, \quad (2.2)$$

and

$$c(1, 1) = \frac{2}{\sqrt{2\pi}} \int_0^{\alpha_1} y \phi(y) h_3(y) dy. \quad (2.3)$$

Let  $B_1(\cdot)$  and  $B_2(\cdot)$  be two independent Gaussian random measures satisfying  $EB_i(A_1)B_i(A_2) = |A_1 \cap A_2|$ ,  $i = 1, 2$ , for all Borel sets  $A_1$  and  $A_2$  of  $\mathbb{R}$ , with  $|\cdot|$  denoting Lebesgue measure. Let  $\Rightarrow$  denote weak convergence in  $C[0, \infty)$ .

**Theorem.** As  $t \rightarrow \infty$ ,

$$Z_t(\tau) = \frac{M(t\tau) - EM(t\tau)}{(\text{Var } M(t))^{1/2}} \Rightarrow \bar{Z}(\tau)$$

in  $C[0, \infty)$ . The limiting process  $\bar{Z}(\tau)$  admits the following Wiener-Itô double integral representation:

$$\begin{aligned} \bar{Z}(\tau) = \frac{K(2, \alpha)}{\sigma} \int_{\mathbb{R}^2} f_\tau(y_1, y_2) \left\{ c(2, 0) \frac{1 + \rho_\infty}{1 + \rho} dB_1(y_1) dB_1(y_2) \right. \\ \left. + c(0, 2) \frac{1 - \rho_\infty}{1 - \rho} dB_2(y_1) dB_2(y_2) \right. \\ \left. + c(1, 1) \left( \frac{1 - \rho_\infty^2}{1 - \rho^2} \right)^{1/2} dB_1(y_1) dB_2(y_2) \right\} \end{aligned}$$

with integrand

$$f_\tau(y_1, y_2) = \int_0^\tau \prod_{i=1}^2 ((s - y_i)^+)^{-\alpha/2 - 1/2} ds, \quad (2.4)$$

and where

$$K(2, \alpha) = \frac{\sqrt{(1 - 2\alpha)(1 - \alpha)}}{\int_0^\infty (x + x^2)^{-\alpha/2 - 1/2} dx} \quad (2.5)$$

and

$$\sigma^2 = 2c^2(2, 0) \left( \frac{1 + \rho_\infty}{1 + \rho} \right)^2 + 2c^2(0, 2) \left( \frac{1 - \rho_\infty}{1 - \rho} \right)^2 + c^2(1, 1) \frac{1 - \rho_\infty^2}{1 - \rho^2}. \quad (2.6)$$

**Warning.** We use the notation  $\int_{\mathbb{R}^2}$  in the Wiener-Itô double integral representation for  $\bar{Z}(\tau)$ , but the definition of that integral can be interpreted as to exclude integration over the diagonal  $y_1 = y_2$  of  $\mathbb{R}^2$ .

**Remarks.** (1) The proof presents an equivalent representation for the limiting process  $\bar{Z}(\tau)$  (see (3.4) below). These representations are valid in the sense of the finite-dimensional distributions. See Fox and Taqqu (1984) for further details about moments or cumulants of processes of the type  $\bar{Z}(\tau)$ .

(2) Maejima (1984) proves the theorem under the assumption  $\tau = 1$ ,  $\rho_\infty = 0$  and  $a = b$ . When  $a = b$  the ellipse is a circle. In that case  $c(1, 1) = 0$  and the cross term vanishes.

(3) Suppose

$$EX_1(s)X_1(s+t) \sim t^{-\alpha_1}L_1(t),$$

$$EX_2(s)X_2(s+t) \sim t^{-\alpha_2}L_2(t),$$

$$EX_1(s)X_2(s+t) \sim EX_2(x)X_1(x+t) \sim t^{-\alpha_3}L_3(t)$$

with  $\alpha_1, \alpha_2, \alpha_3 > 0$  and one of the  $\alpha_i$  is smaller than  $\frac{1}{2}$ . Then the term with the smallest  $\alpha$  will provide the main contribution to the variance and hence will dominate in the limit.

(4) The process  $\bar{Z}(\tau)$  is *non-Gaussian*, is self-similar with index  $H = 1 - \alpha \in (\frac{1}{2}, 1)$ , that is, the finite-dimensional distributions of  $\bar{Z}(a\tau)$  are the same as those of  $a^H \bar{Z}(\tau)$  for all  $a > 0$ . The process  $\bar{Z}(\tau)$  has also stationary increments, and consequently

$$E\bar{Z}(\tau_1)\bar{Z}(\tau_2) = \frac{1}{2}\{\tau_1^{2H} + \tau_2^{2H} - |\tau_1 - \tau_2|^{2H}\}. \quad (2.7)$$

(5) If the domain  $D$  is an arbitrary compact subset in  $\mathbb{R}^2$  in general position, then typically  $c(1, 0)$  and/or  $c(0, 1)$  will be non-zero. In that case the main contributions to the limit will be provided by the term of index  $n = 1$  in the expansion of  $M(t)$  in Hermite polynomials (see (3.2) below). Consequently, and for  $0 < \alpha < 1$ , the limit of  $Z_t(\tau)$  will be the fractional Brownian motion  $B_H(\tau)$  with  $H = 1 - \alpha/2 \in (\frac{1}{2}, 1)$ .  $B_H(\tau)$  is *Gaussian*, has stationary increments, is self-similar with index  $H$  and its covariances are also given by (2.7).

### 3. Proof of the theorem

We shall use the following result whose proof is similar to that of Theorem 6.1 of Fox and Taqqu (1984).

**Proposition 1.** Let  $(W_1(s), W_2(s))$ ,  $s \geq 0$  be a mean zero, stationary Gaussian vector processes with  $EW_1^2(s) = EW_2^2(s) = 1$  satisfying

$$EW_1(s)W_1(s+t) \sim \sigma_1^2 t^{-\alpha}L(t), \quad EW_2(s)W_2(s+t) \sim \sigma_2^2 t^{-\alpha}L(t),$$

$$EW_1(s)W_2(s+t) \sim \kappa\sigma_1\sigma_2 t^{-\alpha}L(t), \quad EW_2(s)W_1(s+t) \sim \kappa\sigma_1\sigma_2 t^{-\alpha}L(t),$$

where  $\sigma_1^2 > 0$ ,  $\sigma_2^2 > 0$ ,  $-1 \leq \kappa \leq 1$ ,  $0 < \alpha < \frac{1}{2}$ , and  $L$  is a slowly varying function. Then

$$\frac{\int_0^t (W_1(s)W_2(s) - EW_1(s)W_2(s)) ds}{t^{1-\alpha}L(t)} \quad (3.1)$$

converges weakly in  $C[0, \infty)$  as  $t$  tends to infinity to

$$\frac{1}{\delta} \int_{\mathbb{R}^2} f_t(y_1, y_2) d\bar{B}_1(y_1) d\bar{B}_2(y_2) \quad (3.2)$$

where  $\bar{B}_1(\cdot)$  and  $\bar{B}_2(\cdot)$  are correlated Gaussian random measures satisfying

$$E\bar{B}_i(A_1)\bar{B}_i(A_2) = \sigma_i^2 |A_1 \cap A_2|, \quad i = 1, 2,$$

$$E\bar{B}_1(A_1)\bar{B}_2(A_2) = \kappa\sigma_1\sigma_2 |A_1 \cap A_2|,$$

for all Borel sets  $A_1$  and  $A_2$ . The constant  $\delta$  equals

$$\delta = \int_0^\infty (x + x^2)^{-\alpha/2-1/2} ds. \quad (3.3)$$

Furthermore, the joint moments of (3.1) converge to those of (3.2) as  $t \rightarrow \infty$ .

**Proof of the theorem.** Proceeding first as in Maejima (1984), we orthogonalize the components of  $X(t) = (X_1(t), X_2(t))$  by setting  $Y(t) = (Y_1(t), Y_2(t)) = (X_1(t), X_2(t))T$  where

$$T = \begin{pmatrix} (2(1+\rho))^{-1/2} & (2(1-\rho))^{-1/2} \\ (2(1+\rho))^{-1/2} & -(2(1-\rho))^{-1/2} \end{pmatrix}$$

so that

$$\tilde{R}(t) \equiv EY(0)^* Y(t) = \begin{pmatrix} \frac{r(t)+\rho(t)}{1+\rho} & 0 \\ 0 & \frac{r(t)-\rho(t)}{1-\rho} \end{pmatrix}.$$

But

$$I\{(X_1(s), X_2(s)) \in D\} = I\{(Y_1(s), Y_2(s)) \in \tilde{D}\}$$

where

$$\tilde{D} = \{(y_1, y_2) : (y_1, y_2)T^{-1} \in D\}.$$

Hence,

$$\begin{aligned} M(t) &= \int_0^t I\{(Y_1(s), Y_2(s)) \in \tilde{D}\} \\ &= \sum_{n=0}^{\infty} \sum_{n_1+n_2=n} c(n_1, n_2) \int_0^t H_{n_1}(Y_1(x)) H_{n_2}(Y_2(x)) ds \end{aligned} \quad (3.4)$$

in  $L^2(\Omega)$  where

$$\begin{aligned} c(n_1, n_2) &= \frac{1}{n_1! n_2!} EI\{(Y_1, Y_2) \in D\} H_{n_1}(Y_1) H_{n_2}(Y_2) \\ &= \frac{1}{n_1! n_2!} \iint_{\tilde{D}} H_{n_1}(y_1) H_{n_2}(y_2) \phi(y_1) \phi(y_2) dy_1 dy_2, \end{aligned}$$

and where the  $H_n$  denote the Hermite polynomials with leading coefficients equal to 1. In particular,  $H_0(y) = 1$ ,  $H_1(y) = y$ ,  $H_2(y) = y^2 - 1$ .

Maejima (1984) proves that  $c(1, 0) = c(0, 1) = 0$ , that  $c(2, 0)$ ,  $c(0, 2)$ ,  $c(1, 1)$  are given by (2.1), (2.2), (2.3) respectively, that  $c(2, 0)$  and  $c(0, 2)$  are non-zero and also that  $Z_t(\tau) = [M(t\tau) - EM(t\tau)]/(\text{Var } M(t))^{1/2}$  has same limit as  $M'(t\tau)/(\text{Var } M'(t))^{1/2}$  where

$$M'(t) = \int_0^t \{c(2, 0)H_2(Y_1(s)) + c(0, 2)H_2(Y_2(s)) + c(1, 1)Y_1(s)Y_2(s)\} ds. \quad (3.5)$$

Note that  $M'(t)$  has mean 0 and the same asymptotic variance as  $M(t)$ . To compute the asymptotic standard deviation of  $M'(t)$ , note that the three terms in the integrand in (3.5) are uncorrelated, and we have

$$EY_1(s)Y_1(s+t) \sim \frac{1+\rho_\infty}{1+\rho} t^{-\alpha} L(t)$$

and

$$EY_2(s)Y_2(s+t) \sim \frac{1-\rho_\infty}{1-\rho} t^{-\alpha} L(t) \quad \text{as } t \rightarrow \infty$$

and

$$EH_2(Y_1(s_1))H_2(Y_1(s_2)) = 2(EY_1(s_1)Y_1(s_2))^2.$$

Therefore

$$(\text{Var } M'(t))^{1/2} \sim \frac{\sigma}{\sqrt{(1-2\alpha)(1-\alpha)}} t^{1-\alpha} L(t)$$

as  $t \rightarrow \infty$ , where  $\sigma$  is defined as in (2.6).

Now set

$$W_1(s) = aY_1(s) + bY_2(s), \quad W_2(s) = cY_1(s) + dY_2(s) \quad (3.6)$$

where  $a, b, c, d$  are constants such that  $W_1(s)W_2(s) - EW_1(s)W_2(s)$  has the same distribution as  $c(2, 0)H_2(Y_1(s)) + c(0, 2)H_2(Y_2(s)) + c(1, 1)Y_1(s)Y_2(s)$ . The constants  $a, b, c, d$  must satisfy the system of equations  $ac = c(2, 0)$ ,  $bd = c(0, 2)$  and  $ad + bc = c(1, 1)$ . Consider for now arbitrary constants  $c(2, 0)$ ,  $c(0, 2)$  and  $c(1, 1)$  satisfying  $4c(2, 0)c(0, 2) \leq c^2(1, 1)$ . This ensures that the constants  $a, b, c, d$  are real-valued. Then

$$EW_1(s)W_1(s+t) \sim \sigma_1^2 t^{-\alpha} L(t),$$

$$EW_2(s)W_2(s+t) \sim \sigma_2^2 t^{-\alpha} L(t),$$

$$EW_1(s)W_2(s+t) \sim EW_2(s)W_1(s+t) \sim \kappa\sigma_1\sigma_2 t^{-\alpha} L(t)$$

where

$$\sigma_1^2 = a^2 \frac{1+\rho_\infty}{1+\rho} + b^2 \frac{1-\rho_\infty}{1-\rho}, \quad \sigma_2^2 = c^2 \frac{1-\rho_\infty}{1-\rho} + d^2 \frac{1-\rho_\infty}{1-\rho},$$

$$\kappa\sigma_1\sigma_2 = ac \frac{1+\rho_\infty}{1+\rho} + bd \frac{1-\rho_\infty}{1-\rho}.$$

Using Proposition 1, we conclude that, as  $t \rightarrow \infty$ ,

$$Z_t(\tau) \Rightarrow \bar{Z}(\tau)$$

weakly in  $C[0, \infty)$  where

$$\bar{Z}(\tau) = \frac{\sqrt{(1-2\alpha)(1-\alpha)}}{\sigma\delta} \int_{\mathbb{R}^2} f_\tau(y_1, y_2) d\bar{B}_1(y_1) d\bar{B}_2(y_2) \quad (3.7)$$

where  $\delta$  is defined in (3.3) and  $\sigma^2$ ,  $\bar{B}_1$ ,  $\bar{B}_2$  are defined in Proposition 1. Note that by (2.5), the multiplicative constant in (3.7) does equal  $K(2, \alpha)/\sigma$ .

Now let  $B_1(\cdot)$  and  $B_2(\cdot)$  be two uncorrelated Gaussian measures satisfying  $EB_i(A_1)B_i(A_2) = |A_1 \cap A_2|$ ,  $i = 1, 2$ . Obtain them by setting

$$\bar{B}_1(A) = a \left( \frac{1+\rho_\infty}{1+\rho} \right)^{1/2} B_1(A) + b \left( \frac{1-\rho_\infty}{1-\rho} \right)^{1/2} B_2(A),$$

$$\bar{B}_2(A) = c \left( \frac{1+\rho_\infty}{1+\rho} \right)^{1/2} B_1(A) + d \left( \frac{1-\rho_\infty}{1-\rho} \right)^{1/2} B_2(A).$$

Then, for arbitrary Borel sets  $A_1$  and  $A_2$ ,

$$\bar{B}_1(A_1)\bar{B}_2(A_2) - E\bar{B}_1(A_1)\bar{B}_2(A_2)$$

has the same distribution as

$$c(2, 0) \left( \frac{1+\rho_\infty}{1+\rho} \right) (B_1(A_1)B_1(A_2) - EB_1(A_1)B_1(A_2))$$

$$+ c(0, 2) \left( \frac{1-\rho_\infty}{1-\rho} \right) (B_2(A_1)B_2(A_2) - EB_2(A_1)B_2(A_2))$$

$$+ c(1, 1) \left( \frac{1-\rho_\infty^2}{1-\rho^2} \right)^{1/2} B_1(A_1)B_2(A_2).$$

Since, for  $A_1$  and  $A_2$  disjoint,

$$E\bar{B}_1(A_1)\bar{B}_2(A_2) = EB_1(A_1)B_1(A_2) = EB_2(A_1)B_2(A_2) = 0,$$

the limiting process  $\bar{Z}(\tau)$  admits the equivalent representation

$$\bar{Z}(\tau) = \frac{K(2, \alpha)}{\sigma} \int_{\mathbb{R}^2} f_\tau(y_1, y_2)$$

$$\times \left\{ c(2, 0) \frac{1+\rho_\infty}{1+\rho} dB_1(y_1) dB_1(y_1) + c(0, 2) \frac{1-\rho_\infty}{1-\rho} dB_2(y_1) dB_2(y_2) \right.$$

$$\left. + c(1, 1) \left( \frac{1-\rho_\infty^2}{1-\rho^2} \right)^{1/2} dB_1(y_1) dB_2(y_2) \right\}. \quad (3.8)$$

This result has been obtained under the assumption that the constants  $c(2, 0)$ ,  $c(0, 2)$  and  $c(1, 1)$  belong to the set

$$S = \{(c(2, 0), c(0, 2), c(1, 1)): 4c(2, 0)c(0, 2) \leq c^2(1, 1)\}.$$

It remains to show that the result holds as well when the constants belong to the complement of  $S$  in  $\mathbb{R}^3$ . Since the moments of  $M'(t\tau)/(\text{Var } M'(t))^{1/2}$  converge to those of  $\bar{Z}(\tau)$  when the constants belong to  $S$ , they will also converge when the constants belong to  $\mathbb{R}^3 \setminus S$  because the convergence of moments is not affected by the values of the constants and because the interior of  $S$  is a non-empty open set. In our framework convergence of moments implies convergence of the finite-dimensional distributions (see for example Fox and Taqqu (1984), proof of Proposition 4.1). Since tightness holds as well, this concludes the proof of the theorem.  $\square$

## References

- [1] R. Fox and M.S. Taqqu, Multiple stochastic integrals with dependent integrators, School of Operations Research and Industrial Engineering Technical Report No. 599, Cornell University, Ithaca, NY (1984). To appear in *Journal of Multivariate Analysis*.
- [2] M. Maejima, Some sojourn time problems for one-dimensional Gaussian processes, Preprint (1984). To appear in *Journal of Multivariate Analysis*.
- [3] M. Maejima, Sojourns of multidimensional Gaussian processes, Preprint (1985). To appear in: Eberlein and Taqqu, eds., *Dependence in Probability and Statistics*, Progress in Probability and Statistics series (Birkhäuser, Boston).